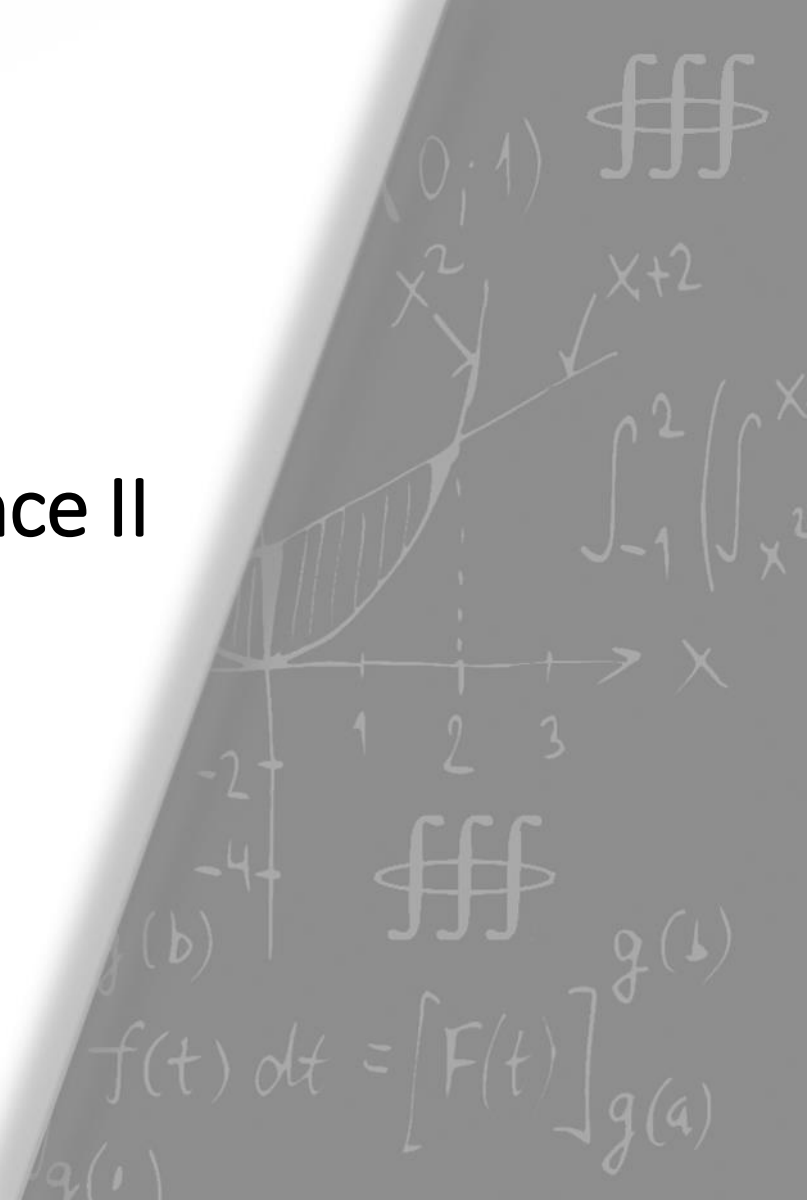




ITRI626 – Artificial Intelligence II

Vaal Triangle



Introduction

- In this session we will dive into what AI is at a high level
- We will look closer at machine learning and then deep learning
- We will identify the main paradigms of machine learning
- We will distinguish between types of learning
- We will introduce neural networks briefly

What is AI?

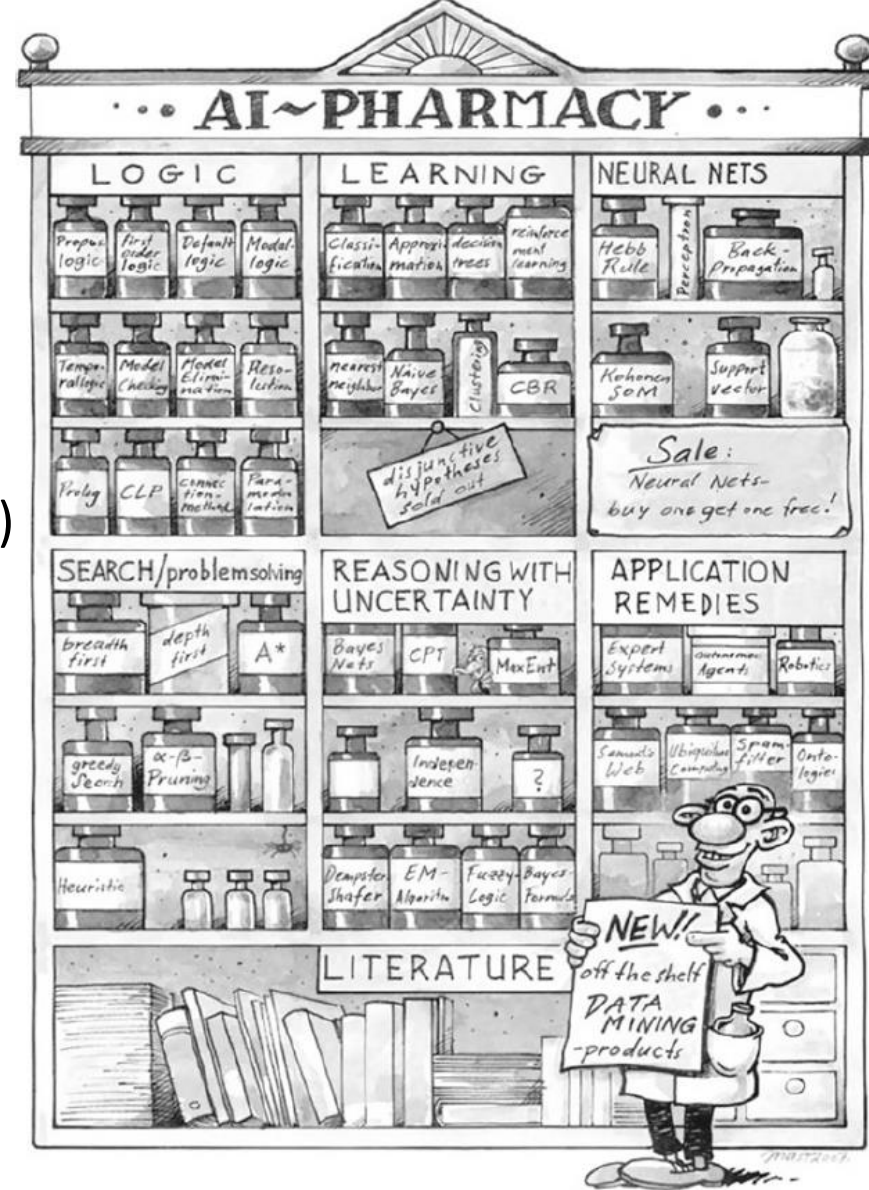
Russel & Norvig:

Thinking Humanly “The exciting new effort to make computers think . . . <i>machines with minds</i> , in the full and literal sense.” (Haugeland, 1985) “[The automation of] activities that we associate with human thinking, activities such as decision-making, problem solving, learning . . .” (Bellman, 1978)	Thinking Rationally “The study of mental faculties through the use of computational models.” (Charniak and McDermott, 1985) “The study of the computations that make it possible to perceive, reason, and act.” (Winston, 1992)
Acting Humanly “The art of creating machines that perform functions that require intelligence when performed by people.” (Kurzweil, 1990) “The study of how to make computers do things at which, at the moment, people are better.” (Rich and Knight, 1991)	Acting Rationally “Computational Intelligence is the study of the design of intelligent agents.” (Poole <i>et al.</i> , 1998) “AI . . . is concerned with intelligent behavior in artifacts.” (Nilsson, 1998)

What is AI?

Ertel – Toolbox:

- Propositional Logic
- Predicate Logic (Application systems)
- Search/problem solving
- Reasoning under uncertainty
- Machine Learning
- Neural Networks



What is AI?

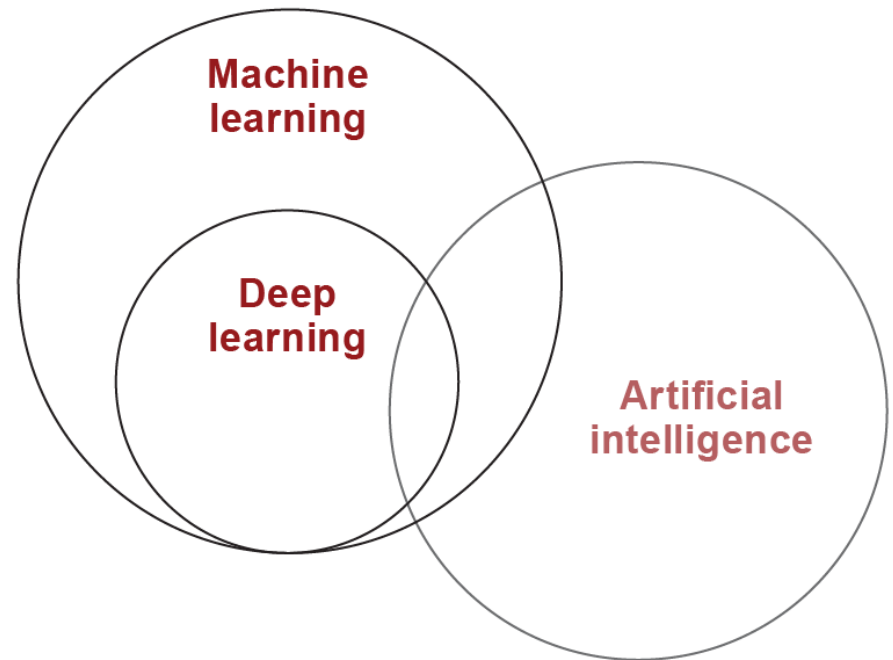
Luger – Application Areas:

- Game playing
- Automated reasoning and Theorem Proving
- Expert Systems
- Natural Language Processing, Understanding and Symantics
- Computer vision
- Modelling human performance
- Planning and robotics
- Languages and environments
- Machine Learning
- Alternative representations: Neural Nets and Genetic Algorithms
- AI and Philosophy

Where does Deep Learning fit?

Task:

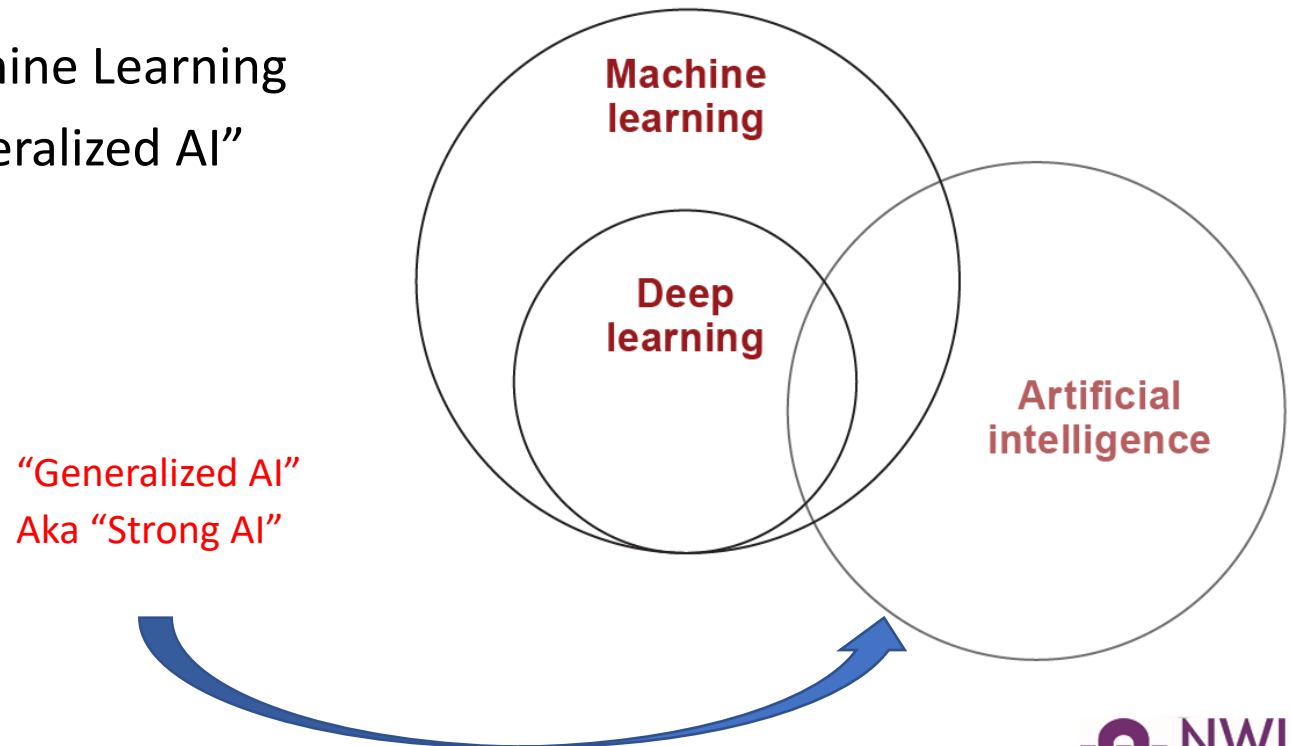
- Subset of Machine Learning



Where does Deep Learning fit?

Task:

- Subset of Machine Learning
- Consider “Generalized AI”



Strong vs Weak AI = General vs Narrow AI

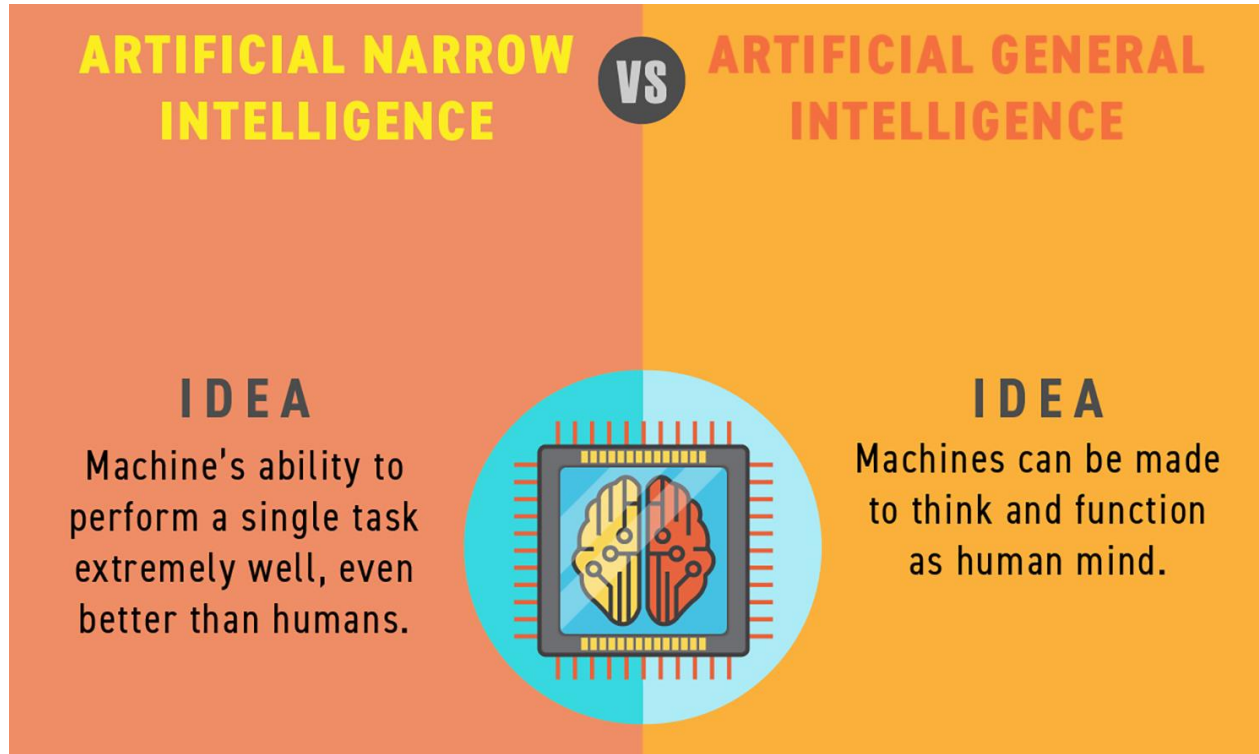
NARROW AI



GENERAL AI



Strong vs Weak AI = General vs Narrow AI



Strong vs Weak AI = General vs Narrow AI

General AI

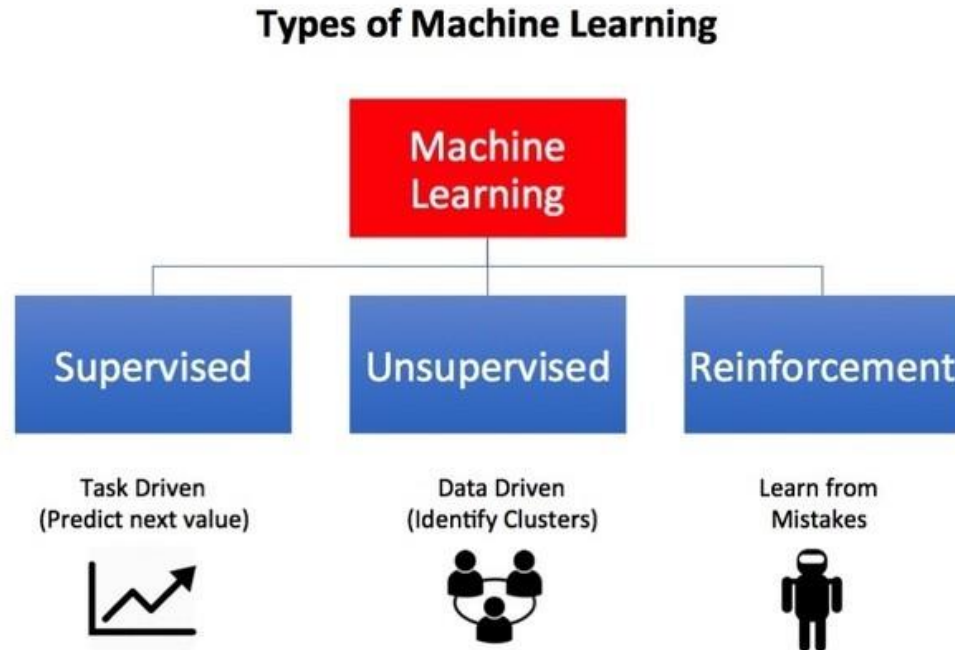
- Driven by scientists
- Multiple tasks
- Understanding

Narrow AI

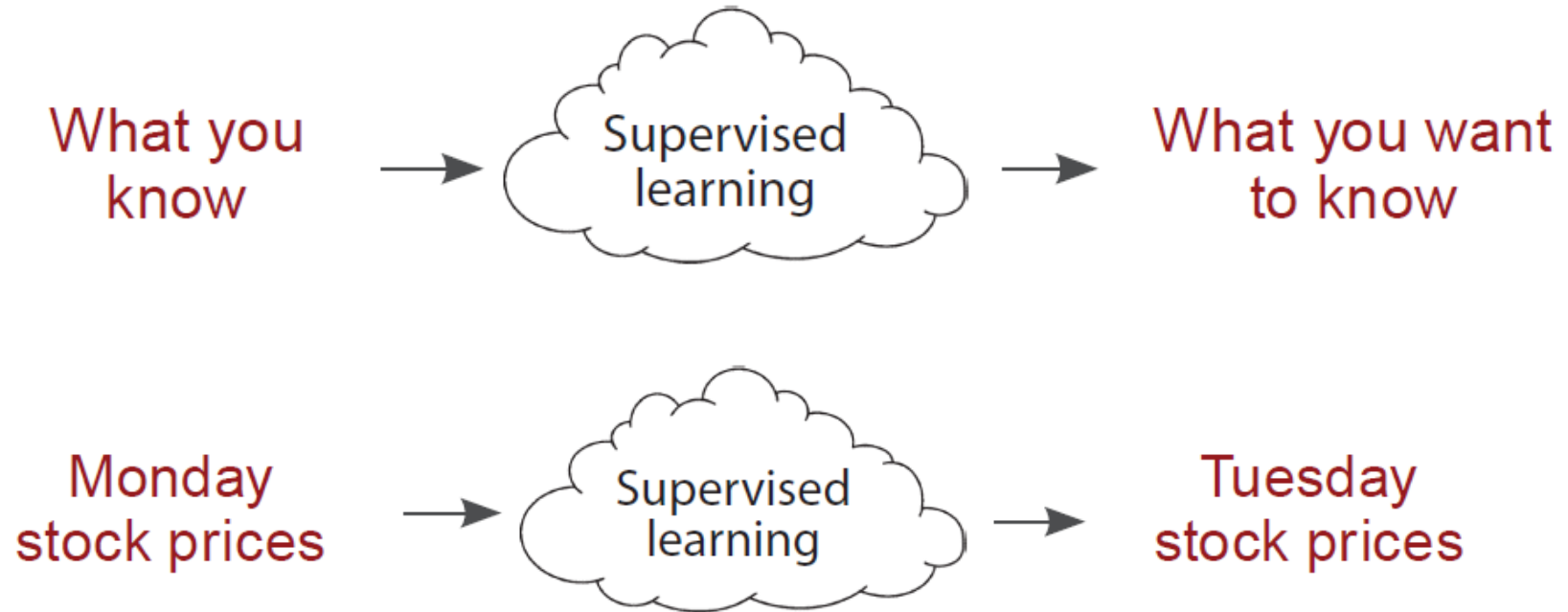
- Driven by Industry
- Single tasks
- Practical

Machine Learning – 3 Paradigms

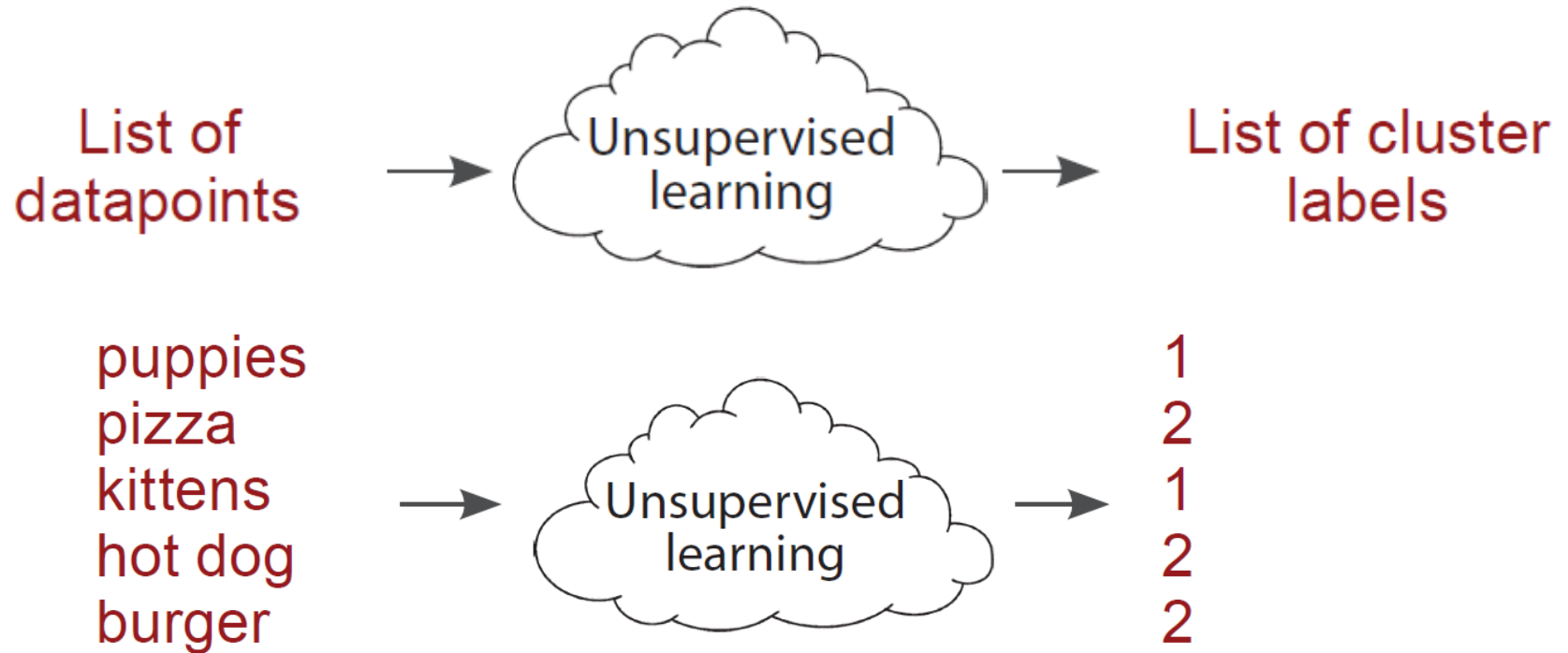
- Machine Learning has 3 main paradigms:



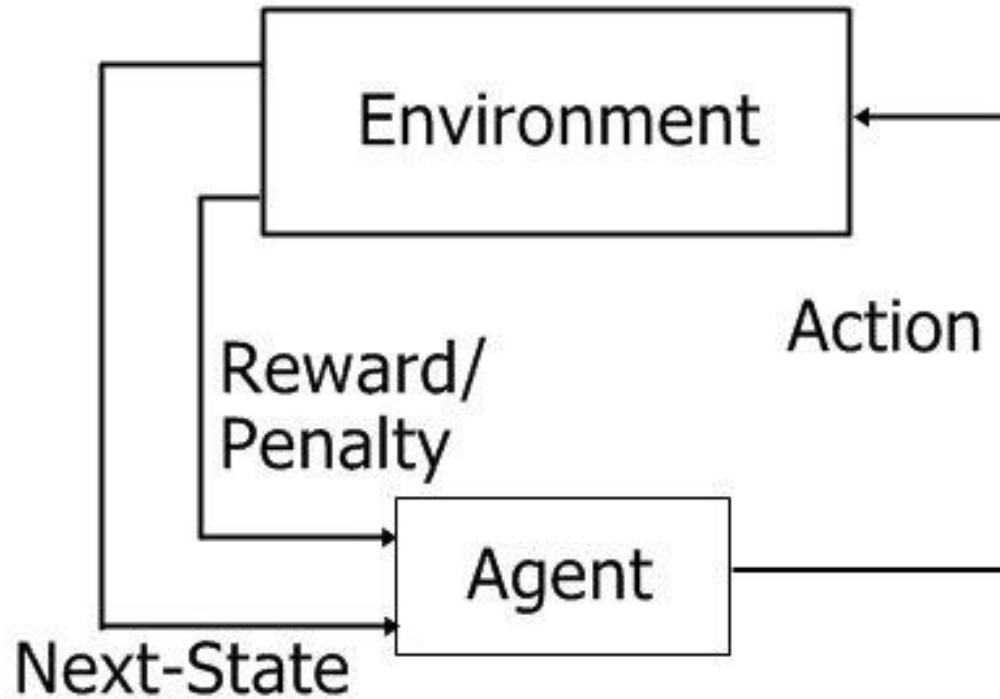
Machine Learning – Supervised Learning



Machine Learning – Unsupervised Learning

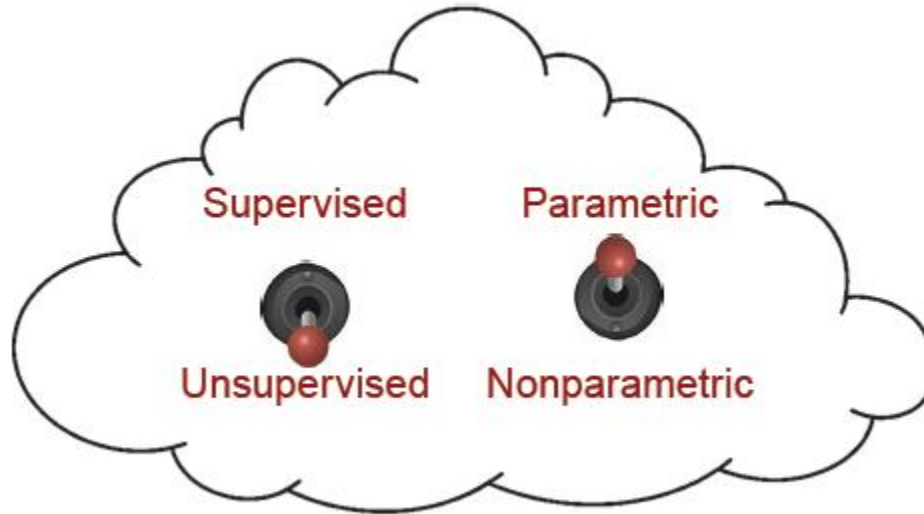


Machine Learning – Reinforcement Learning



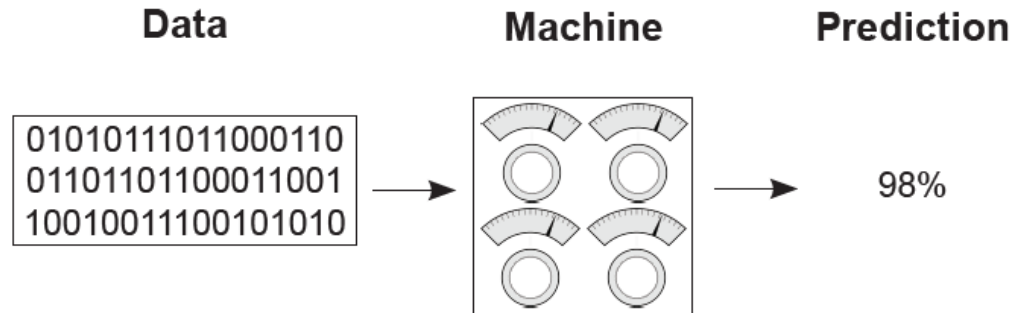
Machine Learning – classifying models

- In addition to the main paradigms, models can also be parametric or non-parametric. (in this course we will focus on supervised and unsupervised learning)



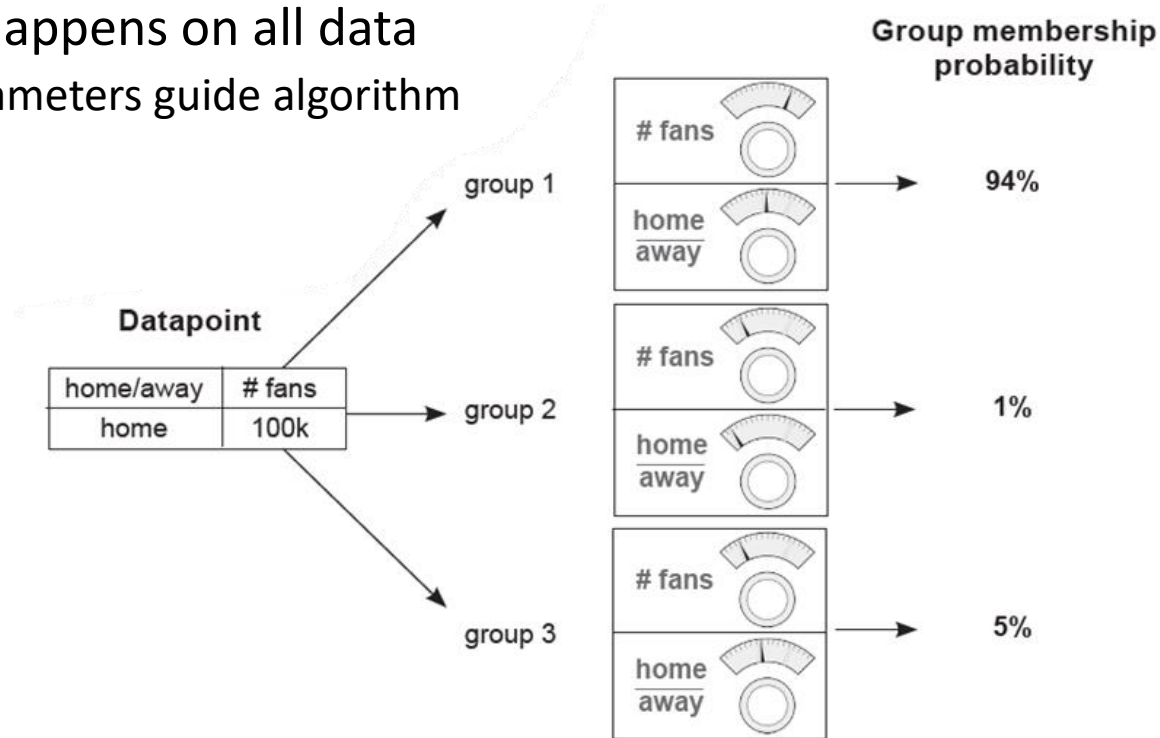
Supervised parametric learning

- Teaching this model type will have 3 steps in a loop:
 - Predict
 - Compare
 - Learn the pattern



Unsupervised parametric learning

- Clustering happens on all data
 - Meta parameters guide algorithm



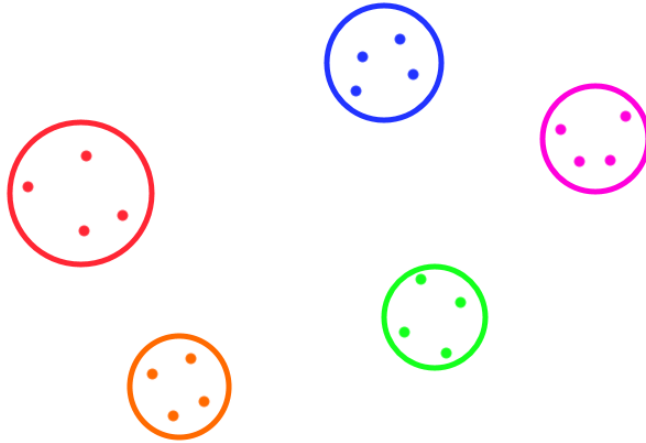
Non-parametric learning

- Data determines the number of parameters, possibly infinite
- Intuitive example of child's sorting cube toy:



Non-parametric learning

- Data determines the number of parameters, possibly infinite
- Example of the k-nearest neighbours algorithm



Python

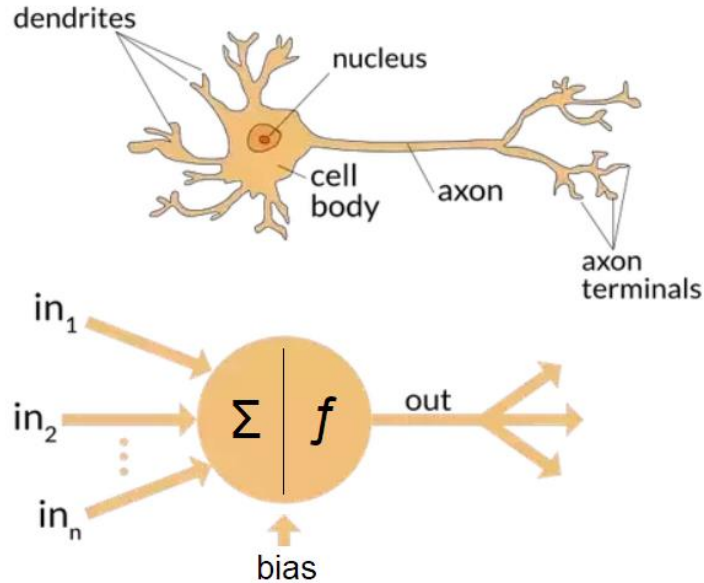
- I would encourage you to use Anaconda
- I would also encourage the use of jupyter notebook (but not prescribed)
- <https://www.edx.org/course/introduction-to-python-absolute-beginner-5>

Important points

- Definitions of AI
- Application areas of AI
- Types of AI
- Machine learning paradigms
- Parametric vs non-parametric learning

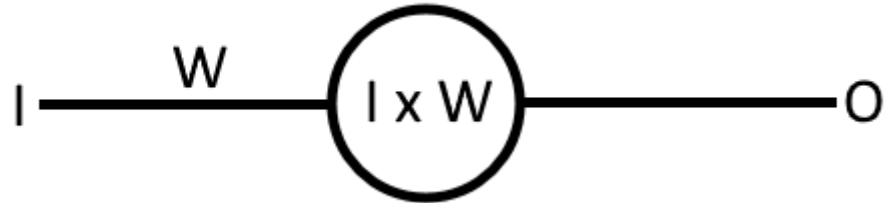
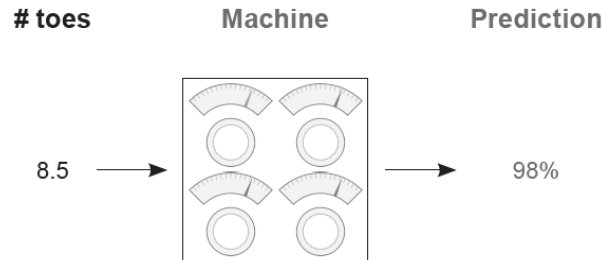
First Neural Network

- Neural networks are modelled on the neurons in the human brain



First Neural Network – SISO

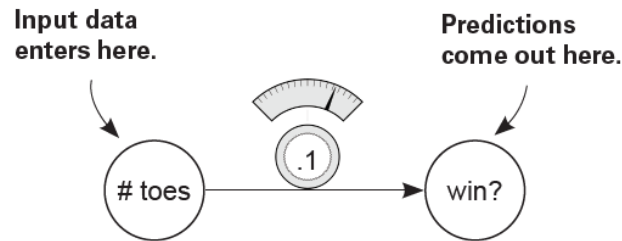
- In a mathematical sense, the first part of the operation that the neural network does is multiply input by some weight that is learned over time to create an output.
- Let us consider then what a single input, single output neural network in its simplest form can look like



*Switch to Python here

First Neural Network - SISO

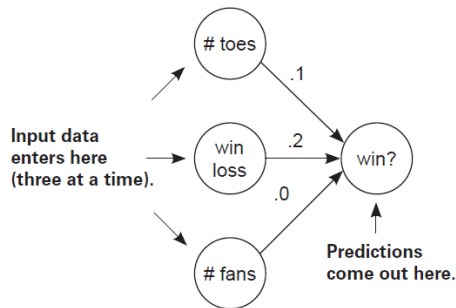
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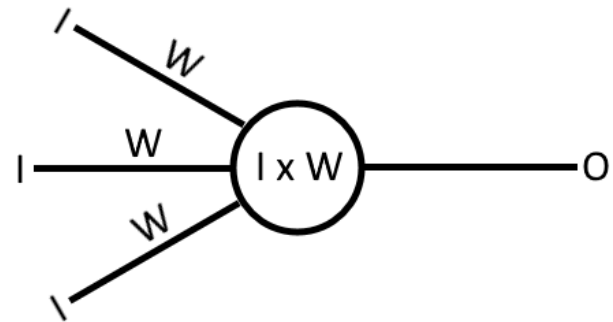
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First Neural Network - MISO

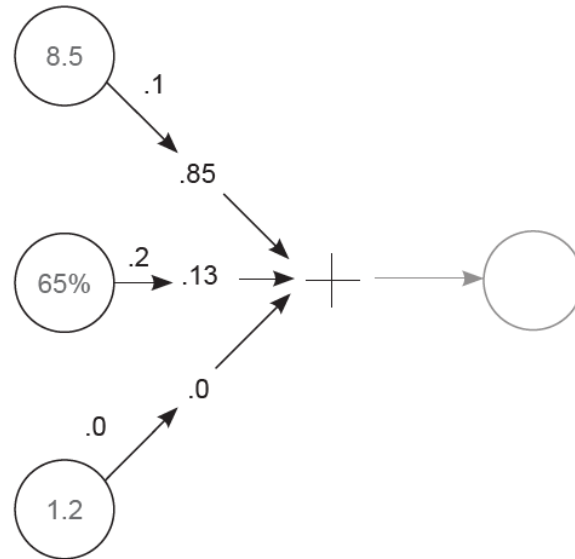
- Next we consider what happens if we have multiple inputs – we will instead take a weighted sum by multiplying each input by its associated weight and summing the result
- Let us consider then what a multi input, single output neural network in its simplest form can look like



*Switch to Python here



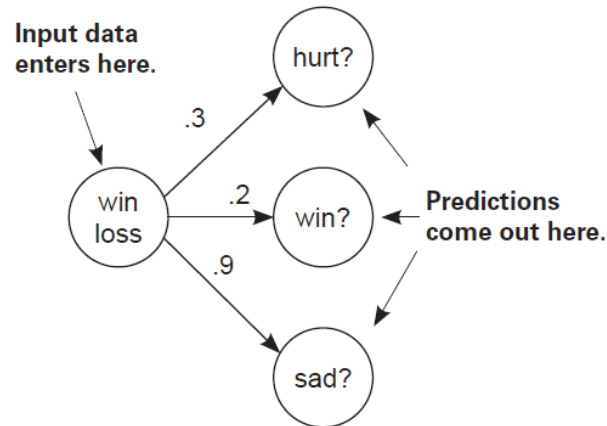
First Neural Network - MISO



Inputs	Weights	Local predictions	
(8.50 * 0.1)	=	0.85	= toes prediction
(0.65 * 0.2)	=	0.13	= wlrec prediction
(1.20 * 0.0)	=	0.00	= fans prediction
toes prediction + wlrec prediction + fans prediction = final prediction			
0.85	+	0.13	+ 0.00 = 0.98

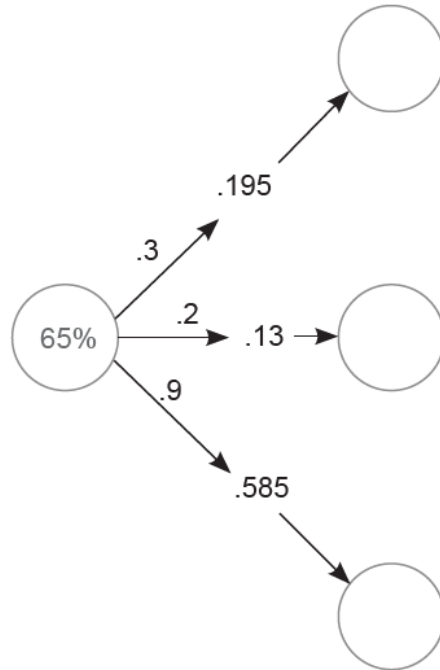
First Neural Network - SIMO

- Multiple outputs are stacked on top of one another to take in the same inputs but use different weights



*Switch to Python here

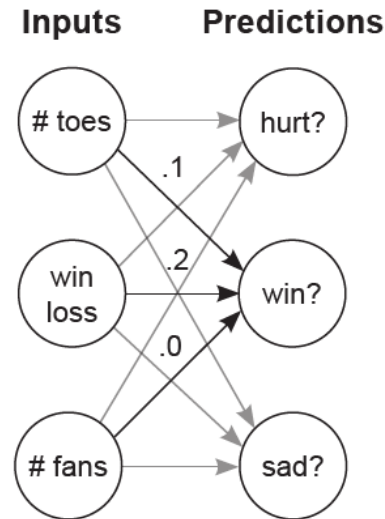
First Neural Network - SIMO



Inputs		Weights		Final predictions	
(0.65	*	0.3)	=	0.195	= hurt prediction
(0.65	*	0.2)	=	0.13	= win prediction
(0.65	*	0.9)	=	0.585	= sad prediction

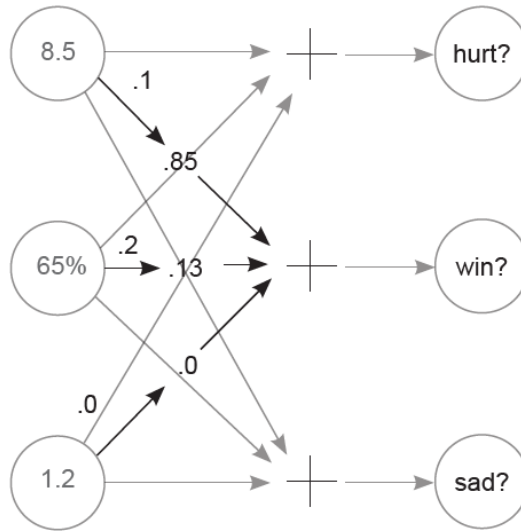
First Neural Network - MIMO

- Multiple outputs are stacked on top of one another to take in the same inputs but use different weights



*Switch to Python here

First Neural Network - MIMO



# toes	% win	# fans			
$(8.5 * 0.1)$	$+$	$(0.65 * 0.1)$	$+$	$(1.2 * -0.3)$	$= 0.555 = \text{hurt prediction}$
$(8.5 * 0.1)$	$+$	$(0.65 * 0.2)$	$+$	$(1.2 * 0.0)$	$= 0.98 = \text{win prediction}$
$(8.5 * 0.0)$	$+$	$(0.65 * 1.3)$	$+$	$(1.2 * 0.1)$	$= 0.965 = \text{sad prediction}$

For Next Week

- Self-study: Chapter 1 (Introducing deep learning) of the Grokking Deep Learning e-book
- 25 March: Practical Quiz 1: Chapter 1 (Introducing deep learning) of the Grokking Deep Learning e-book

References

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